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#### **Examples**

1.we might measure the amount of precipitate P and volume V of gas released from a controlled chemical experiment; we get a two-dimensional sample space consisting of the outcomes (p,v).

2. In a study to determine the likelihood of success in college, based on high school data, one might use a three-dimensional sample space and record for each individual his or her aptitude test score, high school rank in class, and grade-point average at the end of the freshman year in college.

X and Y are two discrete random variables, the **joint** probability distribution of X and Y is

$$p(x,y) = P(X = x, Y = y)$$

the values p(x, y) give the probability that outcomes x and y occur at the same time.

**Definition 3.8** The function p(x, y) is a **joint probability** distribution or **probability mass function** of the discrete random variables X and Y if

- **1.**  $p(x,y) \ge 0$ , for all (x,y),
- **2.**  $\sum_{x} \sum_{y} p(x,y) = 1$ ,
- **3.** P(X = x, Y = y) = p(x, y).

For any region A in the xy plane,

$$P[(X,Y) \in A] = \sum \sum_{A} p(x,y).$$

#### Example 3.8

Two refills for a ballpoint pen are selected at random from a box that contains 3 blue refills, 2 red refills, and 3 green refills. If X is the number of blue refills and Y is the number of red refills selected, find

- (a) the joint probability function p(x,y)
- (b)  $P[(X,Y) \in A]$  where A is the region  $\{(x,y)|x+y \le 1\}$ .

#### Solution

the possible pairs of values (x, y) are (0, 0), (0, 1), (1, 0), (1, 1), (0, 2), and (2, 0).

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p(x,y) represents the probability that  ${\bf x}$  blue and  ${\bf y}$  red refills are selected.

#### Solution

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p(x,y) represents the probability that **x blue** and **y red** refills are selected. The total number of equally likely ways of selecting any 2 refills from the 8 is  $C_8^2$ .

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p(x,y) represents the probability that **x blue** and **y red** refills are selected. The total number of equally likely ways of selecting any 2 refills from the 8 is  $C_8^2$ .

$$p(x,y) = \frac{C_3^x C_2^y C_3^{2-x-y}}{C_8^2}$$

for 
$$x = 0, 1, 2; y = 0, 1, 2; 0 \le x + y \le 2$$
.

#### Solution

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.

present the results in Table 3.1

f(x, y)	x			Row
	0	1	2	totals
10	3 28	9 28	3 28	15 28
y 1	3 14	3 14	20	3 7
2	1/28			$\frac{1}{28}$
lumn totals	5	15 28	3 28	1

(b) 
$$P[(X,Y) \in A] = 9/14$$

the continuous case

**Definition 3.9** The function f(x,y) is a **joint density** function of the continuous random variables X and Y if

- **1.**  $f(x,y) \ge 0$ , for all (x,y),
- **2.**  $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = 1,$
- **3.**  $P[(X,Y) \in A] = \int \int_A f(x,y) dx dy$  for any region A in the xy plane.

When X and Y are continuous random variables, the **joint** density function f(x,y) is a surface lying above the xy plane.

When X and Y are continuous random variables, the **joint** density function f(x,y) is a surface lying above the xy plane.

 $P[(X,Y) \in A]$ , where A is any region in the xy plane, is equal to the volume of the right cylinder bounded by the base A and the surface.

**Example 3.9** Suppose that the joint density function is

$$f(x,y) = \begin{cases} \frac{2}{5}(2x+3y), & 0 \le x \le 1, 0 \le y \le 1\\ 0, & \text{elsewhere,} \end{cases}$$

- (a) Verify condition 2 of Definition 3.9.
- (b) Find  $P[(X,Y) \in A]$ , where A is the region  $\{(x,y)|0 < x < \frac{1}{2}, \frac{1}{4} < y < \frac{1}{2}\}$

Solution.....

**Example 3.9** Suppose that the joint density function is

$$f(x,y) = \begin{cases} \frac{2}{5}(2x+3y), & 0 \le x \le 1, 0 \le y \le 1\\ 0, & \text{elsewhere,} \end{cases}$$

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Solution.....

(b) 
$$P[(X,Y) \in A] = 13/160$$

the discrete case

#### the discrete case

the probability distribution  $p_X(x)$  of X alone is obtained by summing p(x,y) over the values of Y.

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the probability distribution  $p_X(x)$  of X alone is obtained by summing p(x, y) over the values of Y.

Similarly, the probability distribution  $p_Y(y)$  of Y alone is obtained by summing p(x, y) over the values of X.

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 $p_X(x)$  and  $p_Y(y)$ : marginal distributions of X and Y

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 $p_X(x)$  and  $p_Y(y)$ : marginal distributions of X and Y

When X and Y are continuous random variables, summations are replaced by integrals.

**Definition 3.10** The **marginal distribution** of X alone and of Y alone are

$$p_X(x) = \sum_y p(x,y) \quad \text{and} \quad p_Y(y) = \sum_x p(x,y)$$

for the discrete case, and by

$$f_X(x) = \int_{-\infty}^{\infty} f(x,y) dy$$
 and  $f_Y(y) = \int_{-\infty}^{\infty} f(x,y) dx$ 

for the continuous case.

**Example 3.10** Show that the column and row totals of Table 3.1 give the marginal distribution of X alone and of Y alone.

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TABLE 3.1 Joint	Probability Distribution for Example 3.8				
f(x, y)	x			Row -	
	0	1	2	totals	
0	3 28	9 28	3 28	15 28	
y 1	3 14	3 14		3 7	
2	1/28			1/28	
Column totals	<u>5</u>	1 <u>5</u> 28	3 28	1	

**Example 3.11** Find marginal probability density functions  $f_X(x)$  and  $f_Y(y)$  for the joint density function of Example 3.9.

$$f(x,y) = \begin{cases} \frac{2}{5}(2x+3y), & 0 \le x \le 1, 0 \le y \le 1\\ 0, & \text{elsewhere,} \end{cases}$$

# **Example 3.11** Find marginal probability density functions $f_X(x)$ and $f_Y(y)$ for the joint density function of Example 3.9.

$$f(x,y) = \begin{cases} \frac{2}{5}(2x+3y), & 0 \le x \le 1, 0 \le y \le 1\\ 0, & \text{elsewhere,} \end{cases}$$

#### Solution.....

$$f_X(x)=rac{4x+3}{5}$$
 for  $0\leq x\leq 1$ , and  $f_X(x)=0$  elsewhere.  $f_Y(y)=rac{2(1+3y)}{5}$  for  $0\leq y\leq 1$ , and  $f_Y(y)=0$  elsewhere.

The marginal distribution  $p_X(x)$  [or  $f_X(x)$ ] and  $p_X(y)$  [or  $f_Y(y)$ ] are indeed the probability distribution of the individual variable X and Y, respectively.

The marginal distribution  $p_X(x)$  [or  $f_X(x)$ ] and  $p_X(y)$  [or  $f_Y(y)$ ] are indeed the probability distribution of the individual variable X and Y, respectively.

How to verify?

The marginal distribution  $p_X(x)$  [or  $f_X(x)$ ] and  $p_X(y)$  [or  $f_Y(y)$ ] are indeed the probability distribution of the individual variable X and Y, respectively.

How to verify?

The conditions of Definition 3.4 [or Definition 3.6] are satisfied.

recall the definition of conditional probability:

$$P(B|A) = \frac{P(A \cap B)}{P(A)}, \quad P(A) > 0.$$

recall the definition of conditional probability:

$$P(B|A) = \frac{P(A \cap B)}{P(A)}, \quad P(A) > 0.$$

X and Y are discrete random variables, we have

$$P(Y = y | X = x) = \frac{P(X = x, Y = y)}{P(X = x)} = \frac{p(x, y)}{p_X(x)}, \quad p_X(x) > 0.$$

#### **Definition 3.11**

Let X and Y be two discrete random variables. The **conditional probability mass function** of the random variable Y, given that X = x, is

$$p_{Y|X}(y|x) = \frac{p(x,y)}{p_X(x)}, \qquad p_X(x) > 0.$$

Similarly, the conditional probability mass function of the random variable X, given that Y=y, is

$$p_{X|Y}(x|y) = \frac{p(x,y)}{p_Y(y)}, \qquad p_Y(y) > 0.$$

#### Definition 3.11'

Let X and Y be two continuous random variables. The **conditional probability density function** of the random variable Y, given that X = x, is

$$f_{Y|X}(y|x) = \frac{f(x,y)}{f_X(x)}, \qquad f_X(x) > 0.$$

Similarly, the conditional probability density function of the random variable X, given that Y=y, is

$$f_{X|Y}(x|y) = \frac{f(x,y)}{f_Y(y)}, \qquad f_Y(y) > 0.$$

#### Remark:

The function  $f(x,y)/f_X(x)$  is strictly a function of y with x fixed, the function  $f(x,y)/f_Y(y)$  is strictly a function of x with y fixed, both satisfy all the conditions of a probability distribution.

Find the probability that the random variable X falls between a and b when it is known that Y=y

Find the probability that the random variable X falls between a and b when it is known that Y=y

discrete case 
$$P(a < X < b | Y = y) = \sum_{a < x < b} p_{X|Y}(x|y)$$

Find the probability that the random variable X falls between a and b when it is known that Y=y

$$\begin{array}{ll} \text{discrete case} & P(a < X < b | Y = y) = \sum_{a < x < b} p_{X|Y}(x|y) \\ \text{continuous case} & P(a < X < b | Y = y) = \int_a^b f_{X|Y}(x|y) dx. \end{array}$$

**Example 3.12** Referring to Example 3.8, find the conditional distribution of X, given that Y=1, and use it to determine P(X=0|Y=1).

TABLE 3.1 Joint	Probability Distribution for Example 3.8				
		x	Row ·		
f(x, y)	0	1	2	totals	
0	3 28	9 28	3 28	15 28	
y 1	3 14	3 14		3 7	
2	1/28			1/28	
Column totals	5 14	15 28	3 28	1	

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2	1/28			1/28	
Column totals	5 14	15 28	3 28	1	

**Example 3.13** The joint density for the random variables (X,Y) where X is the unit temperature change and Y is the proportion of spectrum shift that a certain atomic particle produces is

$$f(x,y) = \begin{cases} 10xy^2, & 0 < x < y < 1 \\ 0, & \text{elsewhere.} \end{cases}$$

- (a) Find the marginal densities  $f_X(x)$ ,  $f_Y(y)$ , and the conditional density  $f_{Y|X}(y|x)$ .
- (b) Find the probability that the spectrum shifts more than half of the total observations, given the temperature is increased to 0.25 unit.

(a) 
$$f_X(x) = \frac{10}{3}x(1-x^3), \quad 0 < x < 1$$

$$f_Y(y) = 5y^4, \quad 0 < y < 1$$

$$f_{Y|X}(y|x) = \frac{3y^2}{1 - x^3}, \quad 0 < x < y < 1$$

(b) 
$$P(Y > \frac{1}{2}|X = 0.25) = \frac{8}{9}$$

#### Example 3.14 Given the joint density function

$$f(x,y) = \begin{cases} \frac{x(1+3y^2)}{4}, & 0 < x < 2, 0 < y < 1\\ 0, & \text{elsewhere,} \end{cases}$$

find 
$$f_X(x)$$
,  $f_Y(y)$ ,  $f_{X|Y}(x|y)$ , and evaluate  $P(\frac{1}{4} < X < \frac{1}{2}|y=\frac{1}{3})$ .

(a) 
$$f_X(x) = \frac{x}{2}, \quad 0 < x < 2$$

$$f_Y(y) = \frac{1+3y^2}{2}, \quad 0 < y < 1$$

$$f_{X|Y}(x|y) = \frac{x}{2}, \quad 0 < x < 2$$

(b) 
$$P(\frac{1}{4} < X < \frac{1}{2}|Y = \frac{1}{3}) = \frac{3}{64}$$

events A and B are independent, if

$$P(B \cap A) = P(A)P(B).$$

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$$P(B \cap A) = P(A)P(B).$$

discrete random variables X and Y are independent, if

$$P(X = x, Y = y) = P(X = x)P(Y = y)$$

for all (x, y) within their range.

**Definition 3.12** Let X and Y be two discrete random variables, with joint probability distribution p(x,y) and marginal distributions  $p_X(x)$  and  $p_Y(y)$ , respectively. The random variables X and Y are said to be **statistically independent** if and only if

$$p(x,y) = p_X(x)p_Y(y)$$

for all (x, y) within their range.

**Definition 3.12'** Let X and Y be two continuous random variables, with joint probability distribution f(x,y) and marginal distributions  $f_X(x)$  and  $f_Y(y)$ , respectively. The random variables X and Y are said to be **statistically independent** if and only if

$$f(x,y) = f_X(x)f_Y(y)$$

for all (x, y) within their range.

The continuous random variables of Example 3.14 are statistically independent. However, that is not the case for the Example 3.13.

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For discrete variables, requires more thorough investigation.

The continuous random variables of Example 3.14 are statistically independent. However, that is not the case for the Example 3.13.

For discrete variables, requires more thorough investigation. If you find any point (x,y) for which p(x,y) is defined such that  $p(x,y) \neq p_X(x)p_Y(y)$ , the discrete variables X and Y are not statistically independent.

**Example 3.15** Show that the random variables of Example 3.8 are not statistically independent.

f(x, y)	x			Row
	0	1	2	totals
10	3 28	9 28	3 28	15 28
y 1	3 14	3 14	20	3 7
2	1/28			1/28
lumn totals	5	15 28	3 28	1

the case of n random variables

#### the case of n random variables

The marginal distribution for the discrete case is

$$p_{X_1}(x_1) = \sum_{x_2} \cdots \sum_{x_n} p(x_1, x_2, \dots, x_n),$$

and the marginal distribution for the continuous case is

$$f_{X_1}(x_1) = \int_{-\infty}^{\infty} \cdots \int_{-\infty}^{\infty} f(x_1, x_2, \dots, x_n) dx_2 dx_3 \cdots dx_n.$$

joint marginal distributions of two r.v.  $X_1$  and  $X_2$ 

joint marginal distributions of two r.v.  $X_1$  and  $X_2$ 

discrete case

$$p_{X_1,X_2}(x_1,x_2) = \sum_{x_3} \cdots \sum_{x_n} p(x_1,x_2,\ldots,x_n);$$

joint marginal distributions of two r.v.  $X_1$  and  $X_2$ 

discrete case

$$p_{X_1,X_2}(x_1,x_2) = \sum_{x_3} \cdots \sum_{x_n} p(x_1,x_2,\ldots,x_n);$$

continuous case

$$f_{X_1,X_2}(x_1,x_2) = \int_{-\infty}^{\infty} \cdots \int_{-\infty}^{\infty} f(x_1,x_2,\ldots,x_n) dx_3 dx_4 \ldots dx_n.$$

**Definition 3.13** Let  $X_1, X_2, \ldots, X_n$  be n discrete random variables, with joint probability distribution  $p(x_1, x_2, \ldots, x_n)$  and marginal distributions  $p_{X_1}(x_1), p_{X_2}(x_2), \ldots, p_{X_n}(x_n)$ , respectively. The random variables  $X_1, X_2, \ldots, X_n$  are mutually **statistically independent**, then

$$p(x_1, x_2, \dots, x_n) = p_{X_1}(x_1)p_{X_2}(x_2)\cdots p_{X_n}(x_n)$$

for all  $(x_1, x_2, \ldots, x_n)$  within their range.

**Definition 3.13'** Let  $X_1, X_2, \ldots, X_n$  be n continuous random variables, with joint probability distribution  $f(x_1, x_2, \ldots, x_n)$  and marginal distributions  $f_{X_1}(x_1), f_{X_2}(x_2), \ldots, f_{X_n}(x_n)$ , respectively. The random variables  $X_1, X_2, \ldots, X_n$  are mutually **statistically independent**, then

$$f(x_1, x_2, \dots, x_n) = f_{X_1}(x_1) f_{X_2}(x_2) \cdots f_{X_n}(x_n)$$

for all  $(x_1, x_2, \ldots, x_n)$  within their range.

**Example 3.16** Suppose that the shelf life , in years, of a certain perishable food product packaged in cardboard containers is a random variable whose probability density function is given by

$$f(x) = \begin{cases} e^{-x}, & x > 0\\ 0, & \text{elsewhere}, \end{cases}$$

Let  $X_1, X_2$  and  $X_3$  represent the shelf lives for three of these containers selected independently and find

$$P(X_1 < 2, 1 < X_2 < 3, X_3 > 2).$$